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A Novel Bottom-Up Approach to Bounding Low-Dose Human Cancer Risks from Chemical Exposures

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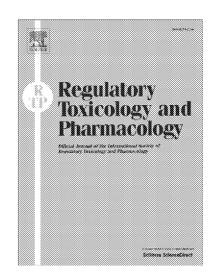
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19	Abstract
20	
21	We propose a novel bottom-up approach to the bounding of low-dose human cancer risks from
22	chemical exposures that does not rely at all upon high-dose data for human or animal cancers.
23	This approach can thus be used to provide an independent "reality check" on low-dose risk
24	estimates derived with dose-response models that are fit to high-dose cancer data. The approach
25	1) is consistent with the "additivity to background" concept, 2) yields central and upper-bound
26	risk estimates that are linear at all doses, and 3) requires only information regarding background
27	risk, background (endogenous) exposure, and the additional exogenous exposure of interest in
28	order to be implemented. After describing the details of this bottom-up approach, we illustrate
29	its application using formaldehyde as an example. Results indicate that recent top-down risk
30	extrapolations from occupational cohort mortality data for workers exposed to formaldehyde are
31	overly conservative by substantial margins.
32	
33	Keywords
34	
35	bottom-up approach, carcinogenic risk assessment, unit risk, q ₁ *, upper 95% confidence bound
36	risk estimate, DNA adducts, formaldehyde
37	
38	1. Background
39	
40	In 1976, Kenny Crump, David Hoel, Charles Langley, and Richard Peto published a landmark
41	paper (Crump et al., 1976) showing that a non-decreasing dose-response relationship for cancer
42	risk will be linear at sufficiently low doses as long as there is a non-zero background exposure to
43	which the specific chemical exposures of interest simply add. This is the well-known "additivity
44	to background" concept: at zero additional exposure, we are already somewhere up on the dose-
45	response curve as a result of our non-zero background exposure, so the slope of the dose-
46	response relationship at zero additional exposure will necessarily be non-zero and positive. Even
47	a threshold dose-response relationship will have a non-zero slope at zero additional exposure if
48	there are some individuals in the population of interest whose thresholds lie below their non-zero

background exposure.

50	Then, in 1977, Crump, Harry Guess, and K.L. Deal published another landmark paper (Crump et
51	al., 1977) that outlined the statistical and mathematical procedures for estimating and bounding
52	the low-dose slope of the multistage dose-response model using constrained maximum
53	likelihood methods and tumor data collected in laboratory animal bioassays conducted at very
54	high exposure levels. It was in this paper that the now infamous " q_1 *", the upper 95%
55	confidence bound on the coefficient of the linear term (i.e., the low-dose slope) of the presumed
56	dose-response relationship, was created, and this value has dominated carcinogenic risk
57	assessment ever since.
58	
59	The dominance of q_1^* in risk assessment has been a consequence of two factors. First, there is
60	the tyranny of small numbers, i.e., the small numbers of animals that have been utilized in
61	laboratory animal carcinogenicity studies, typically, only about 50 animals per sex per dose
62	group. This number is so small that even if the observed tumor incidence in a treated group is
63	zero (0/50), the exact binomial upper 95% confidence limit on the true response probability is
64	0.0582, so true risks up to this value cannot be confidently ruled out. It is also not possible to
65	distinguish statistically at the p=0.05 level between a response as high as 0.08 (4/50 tumor-
66	bearing animals) in a treated group and a null response (0/50) in a control group using Fisher's
67	exact test. If the goal of risk assessment is to bound the dose of a chemical that is associated
68	with an upper bound incremental cancer risk of only 1 per million (10 ⁻⁶), then one can
69	conservatively "guesstimate" the required dose, using the low-dose linear hypothesis, as being
70	about 100,000-fold lower than the highest dose that produces no significant increase, compared
71	to controls, in the probability of developing cancer. This is common knowledge among
72	biostatisticians, and a source of frustration and heartburn among many toxicologists; it is,
73	nevertheless, an irrefutable "fact of life".
74	
75	The second factor behind the dominance of q_1^* is that until recently, the background exposures
76	that may be responsible, at least in part, for our background cancer risks have not been quantified
77	(two notable exceptions are radiation and our background body burdens of dioxin-like
78	compounds). Generally, little attention has been focused on quantifying background chemical
79	exposures, and the exposures of interest have routinely been expressed as increments above
80	whatever the background exposures might be. This is primarily due to the fact that human

background exposures are complicated, uncontrolled, and usually unmeasured, while the animal statement of the statement of th	mal
studies that attempt to carefully control and minimize these background exposures have not	
routinely included measures of the corresponding internal (endogenous) doses that can arise	via
normal metabolism and other internal biochemical reactions.	

Without knowing what background exposure is, expressed preferably as the concentration of a relevant exposure biomarker, e.g., DNA adducts, in the target tissue of interest, the only way to estimate the slope of the dose-response relationship at low doses has been via downward extrapolation from the observed tumor responses in small numbers of animals for occupationally exposed people) at high external exposure levels, which forces us into the q₁*conundrum. However, this situation has changed recently, and the change could profoundly alter carcinogenic risk assessment going forward, at least for those potentially carcinogenic substances that are always present in our bodies, even absent external exposure, because they are produced continuously by normal biochemical processes such as metabolism and biochemical synthesis and degradation. The key technological advance underpinning our novel "bottom-up" approach to risk assessment is the extraordinary ability to distinguish between and separately quantify the relevant internal exposures in target tissues that arise from internal background (endogenous) and external (exogenous) sources. In what follows, we outline this alternative approach to estimating and bounding low-dose cancer risks for such substances, and illustrate the potential for its application with the specific example of formaldehyde, an important commodity chemical that is currently under review by the US Environmental Protection Agency (USEPA).

2. The Bottom-Up Approach

Let P_0 represent the background lifetime risk of a tissue-specific cancer in people, such as nasopharyngeal cancer or leukemia. Let C_0 represent the mean tissue-specific background steady-state concentration of a biomarker, such as a specific DNA adduct, that is presumed to be causally related to these cancers. Then the ratio P_0/C_0 provides an estimate of the low-dose slope of the relationship between the cancer risk and the corresponding tissue-specific DNA adduct concentration. Similarly, if C_{0L} represents the *lower* 95% confidence bound estimate for the same background adduct concentration, then the ratio P_0/C_{0L} provides an *upper* 95% confidence

112	bound on the low-dose slope. This latter ratio is thus directly comparable to the q_1^* derived from
113	high dose animal studies, as well as the upper bound slope estimates for the low-dose linear
114	dose-response relationships that are typically inferred from epidemiologic analyses of
115	occupational cohort cancer mortality, provided only that the dose metrics used in these two kinds
116	of studies (animal bioassays and cohort mortality studies) are converted into the corresponding
117	equivalent tissue-specific adduct concentrations.
118	
119	The key elements of this bottom-up approach are illustrated in Figure 1. What is most important
120	to appreciate is that the central and upper bound slope estimates derived using this approach do
121	not depend in any way on high-dose carcinogenicity data for laboratory animals or humans. The
122	approach thus provides a completely independent "reality check" on low-dose slope estimates
123	like q ₁ * that are derived from analyses of high-dose laboratory animal tumor incidence data or
124	occupational cancer mortality data.
125	
126	3. An Illustration of the Bottom-Up Approach Using Currently Available Data for
127	Formaldehyde
128	
129	Formaldehyde is a highly reactive chemical and an essential metabolic intermediate that is
130	generated endogenously in all living cells, and N ² -hydroxymethyl-deoxyguanosine (dG) adducts
131	have been detected and quantified in various tissues of rats (Lu et al., 2010 and 2011) and
132	cynomolgus macaques (Moeller et al., 2011) exposed to various concentrations of stable isotope-
133	labelled [13CD ₂]-formaldehyde by inhalation. These formaldehyde-DNA adducts are potentially
134	promutagenic because adduction takes place on the amino groups participating in Watson-Crick
135	base pairing, and adduct formation is widely considered to be a key event in the initiation of
136	mutations that lead to carcinogenesis (Swenberg et al., 2011). Thus, the tissue-specific
137	concentration of these adducts provides an excellent internal dose metric with which to illustrate
138	the bottom-up approach to bounding the low-dose slope of dose-response relationships for
139	human cancer risk.
140	
141	The use of [¹³ CD ₂]-formaldehyde permits the simultaneous measurement of both endogenous

143	Ionization-Tandem Mass Spectrometry-Selected Reaction Monitoring (LC-ESI-MS/MS-SRM)
144	methods. While endogenous dG adducts were detected in all of the examined tissues, exogenous
145	dG adducts formed with inhaled [13CD2]-formaldehyde were detected only in the tissues taken
146	from the site of initial contact with exogenous formaldehyde, i.e., rat and monkey nasal
147	respiratory epithelium (Swenberg et al., 2011).
148 149	Because no exogenous dG adducts were detected in these studies in any distant site tissues,
150	including bone marrow and the blood, we can state with confidence that if such exogenous
151	adducts were present in these tissues, then their amounts would necessarily have been smaller
152	than the LC-ESI-MS/MS-SRM method's detection limit (DL). We have therefore used the
153	method's DL (reported in Moeller et al. (2011) as 20 x 10 ⁻¹⁸ mol) as a worst case upper bound on
154	the level of exogenous dG adducts that could be present and yet remain undetected in the bone
155	marrow of $[^{13}CD_2]$ -formaldehyde-exposed monkeys. The above molar DL was converted to an
156	equivalent DL expressed in terms of the number of adducts, namely, 1.03 x 10 ⁻³ per 10 ⁷ dG, using
157	the average amount of monkey DNA collected in the bone marrow samples (Moeller et al.,
158	2011), and the amount of guanine (0.20, expressed as a fraction) that is present in monkey DNA
159	(Casanova et al., 1991).
160	
161	The formaldehyde-DNA adduct data utilized in our bottom-up slope calculations are provided in
162	Table 1. These values are the mean \pm standard error of the number of endogenous and
163	exogenous dG adducts per 10^7 dG in nasal respiratory epithelium (2.49 \pm 0.23 and 0.25 \pm 0.020,
164	respectively), and the bone marrow (17.5 \pm 1.31, endogenous dG adducts only) as determined in
165	monkeys following two 6 hour exposures to 2 ppm [13CD2]-formaldehyde (data taken from Table
166	3 in Swenberg et al., 2011). Also presented are the lower 95% confidence bound estimates for
167	endogenous dG adducts in both tissues, i.e., the mean values minus 1.645 times their respective
168	standard errors.
169	
170	We have also estimated the corresponding steady-state exogenous dG adduct levels that would
171	result from continuous 24 hours/day, 7 days/week exposure. To accomplish this, we used the
172	adduct levels measured in monkeys by Moeller et al. (2011) immediately after the two 6 hour
173	exposures (30 hours after the onset of the first exposure), together with a simple one

174	compartment linear kinetic model of adduct buildup and elimination with a 63 hour elimination
175	half-life (mean adduct lifetime $T = 63/ln(2) = 90.9$ hours) as has been determined in rats
176	(Swenberg et al., in press***). For example, if C_{x30} represents the measured exogenous DNA
177	adduct concentration after two 6 hour exposures on consecutive days to a given airborne
178	formaldehyde concentration, and C_{xS-S} represents the model-predicted asymptotic steady-state
179	adduct concentration that would result from continuous exposure to the same airborne
180	formaldehyde concentration, then $C_{xS-S} = C_{x30}/\{[1-exp(-6/T)]*[1+exp(-24/T)]\} = 8.85 * C_{x30}$.
181	The steady-state adduct concentrations that are predicted by this formula to arise from
182	continuous lifetime exposure to 2 ppm [13CD2]-formaldehyde are also provided in Table 1.
183	
184	At present, we do not have estimates of endogenous or exogenous dG adduct concentrations in
185	human tissues, so we have made the simple assumption that the DNA adduct data collected by
186	Moeller et al. (2011) in cynomolgus macaques are directly relevant to humans without any
187	interspecies scaling adjustments. For the background lifetime risks of developing
188	nasopharyngeal cancer (NPC) and leukemia (LEU), we have relied on two different sources. For
189	NPC, we have taken the estimate of 7.25 x 10 ⁻⁴ that is provided in USEPA's 2 June 2010 draft
190	formaldehyde assessment (see Table C-1, p C-3 and Section 5.2.2). For leukemia, we used the
191	Both Sexes, All Race lifetime risk estimate of 1.3 x 10 ⁻² from Table 1.14 of SEER Cancer
192	Statistics Review 1975-2007 (Altekruse et al., 2010).
193	
194	Table 2 presents the results from using the bottom-up approach with these data and assumptions
195	to calculate upper bound estimates of human nasopharyngeal cancer and leukemia risk from
196	lifetime continuous exposure to 1 ppm formaldehyde. To obtain bottom-up estimates
197	corresponding to 1 ppm formaldehyde, we first calculated bottom-up estimates for 2 ppm (the
198	lowest exposure level used by Moeller et al. (2011), and then simply divided those estimates by a
199	factor of two, since the bottom-up approach assumes linearity of the dose-response relationship.
200	We chose 1 ppm so as to be able to compare our risk estimates simply and directly with those
201	derived by USEPA from epidemiologic data using cumulative formaldehyde exposure as the
202	dose metric, namely 0.011 ppm ⁻¹ for NPC and 0.057 ppm ⁻¹ for leukemia (see Table 6-3, pp 6-41-
203	6-42 of the Agency's draft assessment).

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For nasopharyngeal cancer (NPC), the bottom-up upper bound risk estimate (0.038 x 10 ⁻²) is
nearly 29-fold lower than USEPA's "plausible upper bound" estimate of 1.1 x 10 ⁻² , i.e., about
1%. In contrast, the bottom-up upper bound estimate of leukemia risk (<3.9 x 10 ⁻⁶) is more than
14,000-fold lower than the corresponding USEPA estimate of 5.7 x 10 ⁻² , i.e., about 6%. The
marked disparity between these estimated cancer risks for this distant site suggests strongly that
the excess risk of leukemia that has been reported in association with workplace formaldehyde
exposures is not due to those exposures. If our plausible assumption that formaldehyde dG
adducts provide a valid molecular dosimeter for relating potential human cancer risks to
formaldehyde exposure is correct, then the much larger risks derived by USEPA from the adult
human cancer data are overly conservative.
4. Strengths and Limitations of the Bottom-Up Approach

We are confident that the estimates obtained from this simple approach to bounding low-dose human cancer risks are conservative for several reasons. Most importantly, the bottom-up approach attributes all of the background risk of specific cancers to the endogenous formaldehyde dG adducts that are found in the corresponding tissues. If only a fraction f of the total background risk P₀ were due to the background endogenous adduct concentration C₀, then it is only this fraction of the total background risk P₀ to which our assumed linear dose-response relationship should apply. The slope estimate P_0/C_0 and the associated upper 95% confidence bound described herein would therefore exaggerate the actual slope of the low-dose response and its upper estimated 95% confidence bound by the factor 1/f.

Second, the approach is linear at low doses simply because it assumes linearity at all doses. This can create problems if we extrapolate the bounding bottom-up risk estimates to very high exogenous exposure levels, such as those producing statistically significant increases in tumor incidence in exposed laboratory animals or humans. At such levels, it is expected that the doseresponse relationship for tumor incidence may well be highly non-linear due to a variety of factors that become important only at high doses, such as cytotoxicity, tissue damage, and enhanced cell proliferation that markedly increases the probability of mutations. Such nongenotoxic high dose phenomena are not accounted for in our simple linear model, so the

236	confidence bounds that it generates should not be expected to hold at the high exogenous
237	exposures where these phenomena take place.
238	
239	Third, we have used <i>lower</i> 95% confidence bounds on the estimated mean endogenous DNA
240	adduct levels (C ₀) to generate, by simple inversion, the corresponding <i>upper</i> 95% confidence
241	bounds on the slope (P_0/C_0) of the linear relationship that has been assumed between cancer risks
242	and steady-state adduct concentrations. This follows directly from a Taylor series expansion of
243	the ratio P_0/C_0 about its expected value when P_0 is taken to be constant, as we have assumed
244	herein, and only the estimated mean background DNA adduct concentration Co has uncertainty
245	associated with it. In this special case, the variance of the ratio P_0/C_0 is given approximately by
246	(c.f., Stuart and Arnold, 1994, p. 351):
247	
248	$Var(P_0/C_0) / (P_0/C_0)^2 \approx Var(C_0) / {C_0}^2$. Equation 1
249	
250	If uncertainty in the estimate of the mean background risk P ₀ is also characterized and it is
251	independent of, i.e., uncorrelated with, the uncertainty in C ₀ , as is the case herein, because the
252	estimates of P ₀ and C ₀ are derived from two completely different data sets, then the additional
253	uncertainty in the estimated ratio P_0/C_0 that is due to the uncertainty in P_0 can also be readily
254	accommodated (ibid.):
255	
256	$Var(P_0/C_0) / (P_0/C_0)^2 \approx Var(C_0) / C_0^2 + Var(P_0) / P_0^2$. Equation 2
257	
258	The upper 95% confidence bound on the slope estimate P_0/C_0 is thus increased as a result of
259	incorporating the additional uncertainty in P ₀ . Additional discussion of uncertainty in the
260	estimate of the ratio P_0/C_0 in a linear regression context is provided in the Appendix.
261	
262	Fourth, for the case of leukemia, exogenous DNA adducts were never detected in monkey bone
263	marrow even though the methodology has the sufficient statistical power to detect a single N ² -
264	hydroxymethyl-deoxyguanosine adduct in 10 billion deoxyguanosine molecules. We have
265	therefore assumed, as a worst case, that exogenous DNA adducts could have been present at a

level just barely below the detection limit of the ultrasensitive LC-ESI-MS/MS-SRM

267	methodology. One could reduce this estimate substantially using less conservative assumptions
268	regarding the sampling distribution of non-detected exogenous DNA adduct concentrations (c.f.,
269	Ginevan and Splitstone, 2004, pp 123-125).
270	
271	Fifth, we have made reasonable assumptions in converting adduct concentrations measured in
272	monkey tissues after two 6 hour/day exposures to 2 ppm airborne formaldehyde on consecutive
273	days to the higher steady-state adduct levels that would arise from continuous exposures to the
274	same airborne formaldehyde concentration for a lifetime, but our extrapolation using a simple
275	linear pharmacokinetic model has not yet been validated. Data from longer term studies out to
276	28 consecutive days of exposure are currently being analyzed, so the remaining uncertainty
277	regarding the half-life of formaldehyde DNA adducts should be better resolved in due course.
278	
279	Even so, the cross-species extrapolation from DNA adduct data obtained in monkeys to human
280	formaldehyde exposures remains unvalidated. However, unvalidated assumptions can be
281	replaced at some point with data-driven alternatives. For example, in the near future, we expect
282	to obtain data regarding endogenous formaldehyde dG adducts in human tissues. Human blood
283	samples are readily obtainable, and opportunistic sampling of other critical tissues such as nasal
284	tissue and bone marrow is certainly possible. Such data could be used to confirm and/or replace
285	our plausible dosimetric assumption that endogenous formaldehyde dG adduct amounts in
286	monkey and human tissues are directly comparable. Obtaining exogenous adduct concentrations
287	in humans may be more problematic. However, the extraordinary sensitivity of the Lu et al.
288	(2010, 2011) and Moeller et al. (2011) methodology may offer the prospect of detecting such
289	adducts using short-term voluntary human exposures.
290	
291	Finally, another important limitation of the bottom-up approach is its reliance on the assumption
292	of a linear dose-response relationship between cancer risks and DNA adduct concentrations in
293	target tissues, even though the true dose-response may be highly nonlinear at sufficiently high
294	exogenous exposure levels. For this reason, we advocate the bottom-up approach only as a
295	potential means for generating tighter upper bounds on low-dose human cancer risks than it may
296	be possible to achieve with top-down approaches. The bottom-up approach may not be
297	appropriate for developing "best" or central estimates of low-dose human cancer risks which, at

298	least in our view, can best be accomplished through a comprehensive and deep mechanistic
299	understanding of exactly how chemical exposures give rise to human cancer.
300	
301	5. Summary
302	
303	The Lu et al. (2010, 2011) and Moeller et al. (2011) LC-ESI-MS/MS-SRM methodology
304	differentiates clearly between DNA adducts formed with formaldehyde molecules of endogenous
305	and exogenous (inhaled) origin. This remarkable technological achievement has made it
306	possible to develop upper-bound estimates of potential cancer risk with a unique bottom-up
307	approach that extrapolates upward from background (endogenous) exposures and background
308	cancer risks, as opposed to the typical top-down extrapolations from cancer incidence in
309	laboratory animals or human workers subjected to very high exposure levels.
310	
311	While we have illustrated the bottom-up approach with the example of formaldehyde, we expect
312	it to be readily generalizable to other chemicals. For example, vinyl chloride, ethylene oxide,
313	methanol, and acetaldehyde are all known to produce specific DNA adducts from endogenous
314	and exogenous sources, and other chemicals are likely to be added to this list in the near future.
315	The target tissue dose concept can also be generalized to include other forms of endogenous
316	DNA damage, such as abasic sites, lesions arising from oxidative stress, and also biomarkers of
317	effect, such as mutations (Swenberg et al., 2008 and 2011). The potential of the bottom-up
318	approach to impact human cancer risk assessment appears great.
319	
320	We used the new molecular dosimetry information for formaldehyde DNA adducts in the
321	bottom-up approach to estimate upper-bound lifetime human nasopharyngeal cancer and
322	leukemia risks that might arise from continuous inhalation exposure to 1 ppm formaldehyde.
323	This provides a totally independent "reality check" on estimates derived with the conventional
324	top-down approach to human cancer risk assessment. Comparison of the resulting bottom-up
325	risk estimates with corresponding top-down estimates derived by USEPA from epidemiologic
326	data for exposed workers show the latter to be markedly higher. The large discrepancies
327	between the results we obtained with molecular dosimetry data incorporated into the bottom-up
328	approach and those that relied on worker cancer mortality and uncertain retrospective

329	occupational exposure reconstructions call into serious question the credibility of attributing
330	large increases in human mortality from these cancers to occupational formaldehyde exposure.
331	
332	6. Acknowledgements
333	
334	TBS received financial support for this work from the Research Foundation for Health and
335	Environmental Effects, Washington DC. We also thank an anonymous reviewer for pointing out
336	that estimating an upper bound on the slope of the dose-response relationship at and near the
337	background exposure level can be considered in the context of linear regression through the
338	origin.
339	
340	7. Conflict of Interest Statement
341	
342	TBS has served as a consultant on risk assessment issues related to formaldehyde for the
343	American Chemistry Council. The formaldehyde research conducted by JAS has been funded in
344	part by the NIEHS, the American Chemistry Council, Formacare, and the Texas Commission for
345	Environmental Quality. The sponsors do not have access to research results until they have been
346	accepted for publication. JAS has also served as a formaldehyde consultant to ENVIRON
347	International.

348		
349	8.	References
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404				
405	Appendix			
406				
407	An anonymous reviewer pointed out to us that estimation of an upper percentile confidence			
408	bound on the slope of the dose-response curve near the background exposure level C ₀ can also be			
409	considered in the context of linear regression, with the regression line forced through the origin			
410	(c.f., Neter et al., 1996, pp 159-163). The situation we are concerned with here is a special one,			
411	because there is only one Y value, namely P ₀ , which is derived from national cancer statistics,			
412	i.e., the $Y_i\left(P_{0i}\right)$ values that would be associated in a regression context with the n individual X_i			
413	measurements of the background endogenous DNA adduct concentration (C_{0i}) are actually all			
414	equal to the single mean background risk estimate P ₀ . Given this constraint, it is not difficult to			
415	show algebraically that the slope estimate b ₁ resulting from linear regression through the origin			
416	of the equal Y_i on the individual C_{0i} is given by:			
417				
418	$b_1 = P_0/C_0 * \{1 + ((n-1)/n)*Var(C_0)/C_0^2\},$ Equation A1			
419				
420	where $Var(C_0)$ is the estimated variance of the C_{0i} . So long as $Var(C_0)$ is small compared to ${C_0}^2$,			
421	the term in curly brackets in Equation A1 will be close to unity in value, and the regression slope			
422	estimate b_1 will be approximately equal to P_0/C_0 , but in any case, b_1 will always be <i>smaller</i> than			
423	the ratio P_0/C_0 . Thus P_0/C_0 will be a conservative, i.e., larger, estimate of the low-dose slope			
424	than the linear regression estimate b ₁ given in Equation A1.			

14

126	In addition, the estimated variance of the regression slope estimate b ₁ for this special case of			
427	equal Yi can be shown algebraically to be given by:			
428				
129	$Var(b_1) = b_1^2 * [(Var(C_0)/n)/C_0^2].$ Equation A2			
430				
431	Thus, the percentiles of the sampling distribution for b ₁ in this special case are exactly			
432	complementary to those of the sampling distribution of the mean background DNA adduct			
433	concentration C_0 , i.e., an upper 95% confidence bound on the regression slope estimate b_1			
134	corresponds exactly to a lower 95% confidence bound on the mean background DNA adduct			
435	concentration.			
436				
437	When the additional uncertainty associated with individual background risk estimates P_{0i} is			
438	considered, so long as there is no covariance between the P_{0i} and C_{0i} , as is the case here, the			
139	regression estimate of b ₁ remains the same as that given by equation A1. However, the estimated			
440	variance of b ₁ changes to:			
441				
142	$\operatorname{Var}(b_1)/b_1^2 = (\operatorname{Var}(C_0)/n)/{C_0}^2 + (\operatorname{Var}(P_0)/n)/{P_0}^2) * [1 + ((n-1)/n)*\operatorname{Var}(C_0)/{C_0}^2],$			
143	Equation A3			
144				
145	where $Var(P_0)$ represents the variance of the individual background risk measurements P_{0i} .			
146	Thus, the upper 95% confidence bound estimate for the regression slope estimate b ₁ is increased			
147	as a result of incorporating the additional uncertainty associated with the individual background			
148	risk estimates P _{0i} .			

449	Highlights
450	The new bottom-up approach does not rely on high-dose cancer data to estimate low-dose
451	risk
452	
453	It utilizes background human cancer risk and background exposures from endogenous
454	sources
455	
456	It is low-dose linear and consistent with the "additivity to background" risk concept
457	
458	It provides an independent "reality check" on top-down estimates derived from high-dose
459	animal or human cancer data
460	
461	For formaldehyde, the bottom-up risk estimates are markedly lower than recent USEPA
462	top-down estimates
463	
	y .

Figure Captions
Figure 1. Illustrating the "bottom-up" approach to bounding additional human cancer risks that
may be associated with low level chemical exposures. Po is the background lifetime risk of a
tissue-specific cancer. C ₀ and C _{0L} are the central and lower 95% confidence bound estimates of
the steady-state background concentration of specific DNA adducts linked to the cancer in the
same tissue. β and β_u are the bottom-up central and upper 95% confidence bound estimates of
the low-dose slope of the cancer risk-DNA adduct relationship.

473 F1: 474

475

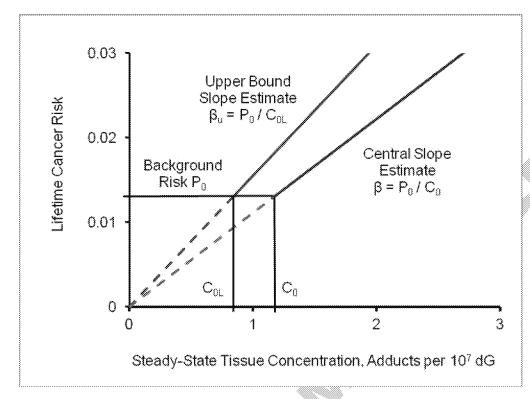


Table 1. Endogenous and exogenous DNA adduct concentrations (per $10^7 dG$) in nasal epithelial tissue and bone marrow of cynomolgus macaques exposed via inhalation for 6 hours on two consecutive days to 2 ppm [$^{13}CD_2$]-formaldehyde (data taken from Moeller et al. (2011)). Also shown are the 8.85-fold higher steady-state exogenous adduct concentrations that are expected to result from lifetime continuous inhalation exposure to 2 ppm [$^{13}CD_2$]-formaldehyde (see text for details).

Tissue	Endogenous Adducts at 30 hrs	Exogenous Adducts at 30 hrs	Exogenous Adducts at Steady-State
Nasal Epithelium Mean ± se Lower 95% Bound	2.49 ± 0.23 2.11	0.250 ± 0.020	2.21 ± 0.18
Bone Marrow Mean ± se Lower 95% Bound	17.5 ± 1.31 15.34	< 0.00103 ^a	<0.00912 ^a

 a: no exogenous adducts were detected in bone marrow; upper limits estimate based on the detection limit reported in Moeller et al. (2011).



Table 2. Comparison of estimated lifetime risks of developing nasopharyngeal cancer (NPC) and leukemia (LEU) from continuous lifetime inhalation exposure to 1 ppm formaldehyde, as estimated with the bottom-up approach and, alternatively, by USEPA using top-down linear extrapolation from epidemiologic data (as taken from Table 6-3, pp 6-41-6-42 of the Agency's 2 June 2010 draft assessment).

Cancer	Background Risk, P ₀	Bottom-Up Slope, P ₀ /C _{0L} ^a	Bottom-Up Risk at 1 ppm ^b	USEPA Risk at 1 ppm
NPC	7.25 x 10 ⁻⁴	3.44 x 10 ⁻⁴	0.038 x 10 ⁻²	1.1 x 10 ⁻²
LEU	1.30 x 10 ⁻²	8.50 x 10 ⁻⁴	< 3.9 x 10 ⁻⁶	5.7 x 10 ⁻²

a: for NPC,
$$3.44 \times 10^{-4} = 7.25 \times 10^{-4} / 2.11$$

for LEU, $8.50 \times 10^{-4} = 1.30 \times 10^{-2} / 15.3$

b: for NPC,
$$0.038 \times 10^{-2} = 3.44 \times 10^{-4} \times (2.21/2)$$

for LEU, $< 3.9 \times 10^{-6} = 8.50 \times 10^{-4} \times (< 0.00912/2)$